
Table of Contents

.....	1
Creating simulated data	1
Setting up initial samples	1
Sampling θ using Hamiltonian Monte Carlo	1
Sampling (θ, α_t) using Hamiltonian Monte Carlo	3
Sampling (θ, α_t) using T-factors Auxiliary Gibbs sampler	7
Sampling θ using adaptive Metropolis sampler	11
Cleaning up working folder	12

```
% This script generates the posterior samples for the model
%
% $$ \alpha_t \sim N(0,1); t=1,2,\dots,T $$
%
% $$ y_t \sim N(\alpha_t, \exp(\theta)) $$
%
```

Creating simulated data

```
clc; close all
T      = 100;                % data points
alpha  = randn(1,T);
y      = alpha + randn(1,T);
```

Setting up initial samples

```
theta  = 0;    % first parameter
alpha_T = y./2; % mean value of the Kalman filter output
```

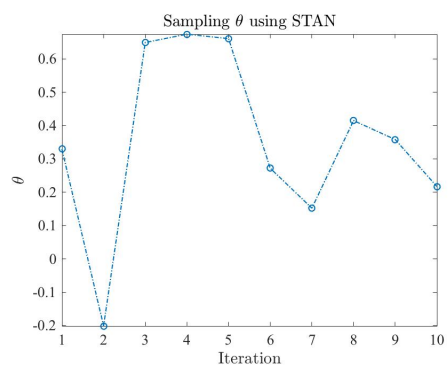
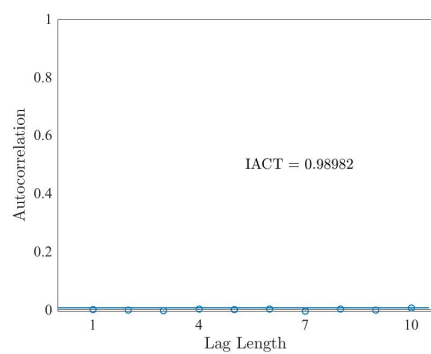
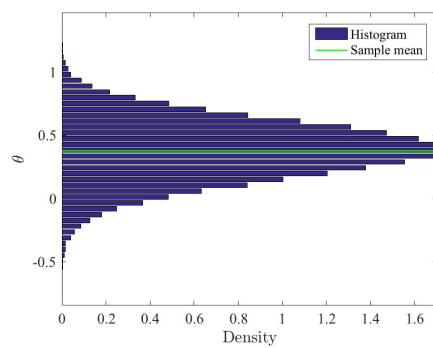
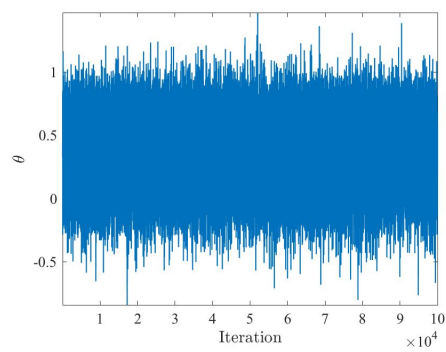
Sampling θ using Hamiltonian Monte Carlo

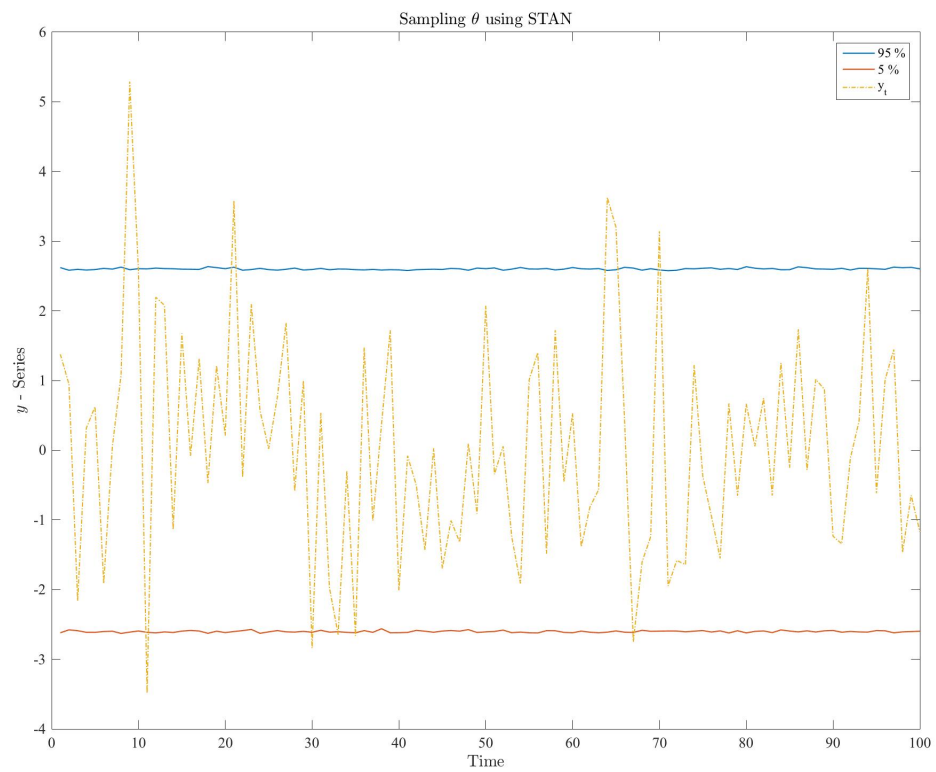
```
tic
launch_stan
time_stan = toc
```

Stan succeeded.

```
time_stan =

    36.0408
```



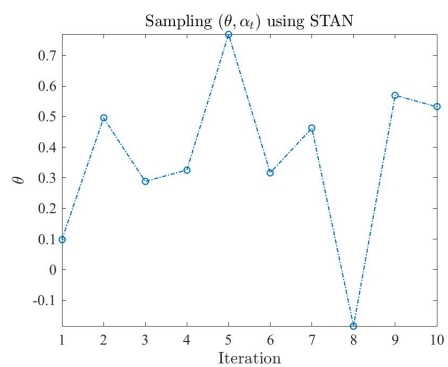
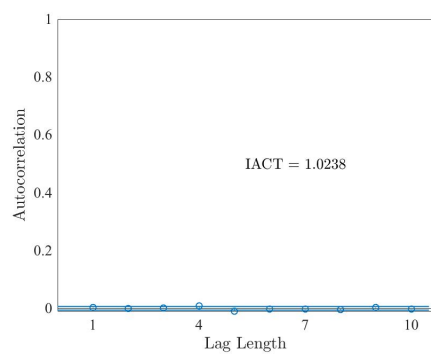
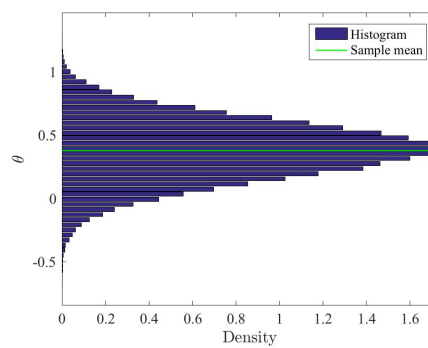
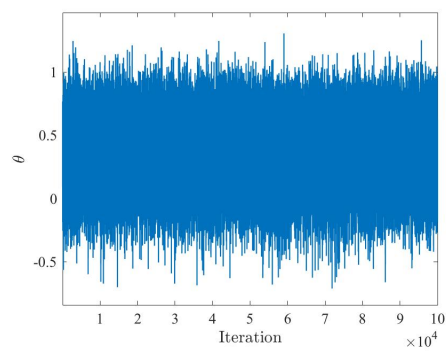


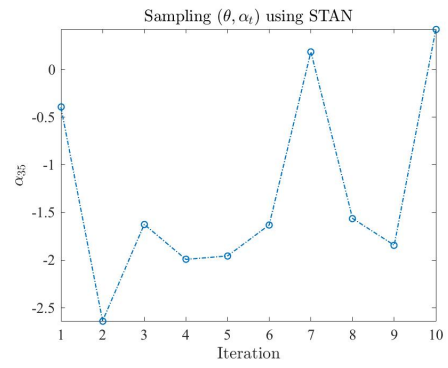
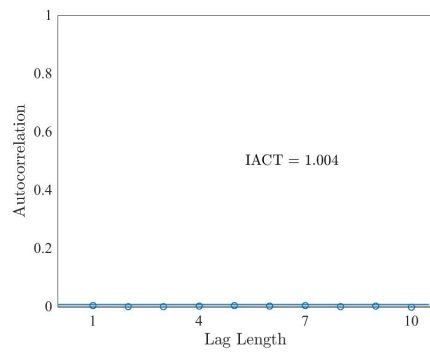
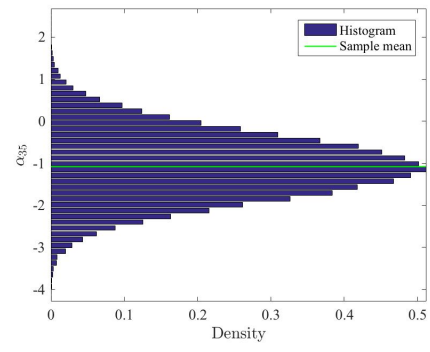
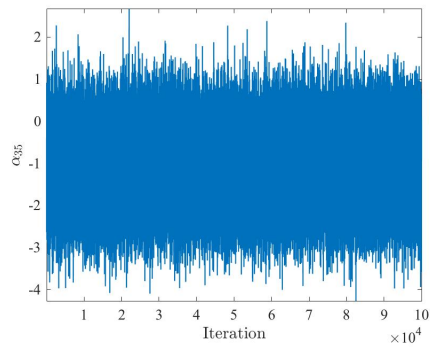
Sampling (θ, α_t) using Hamiltonian Monte Carlo

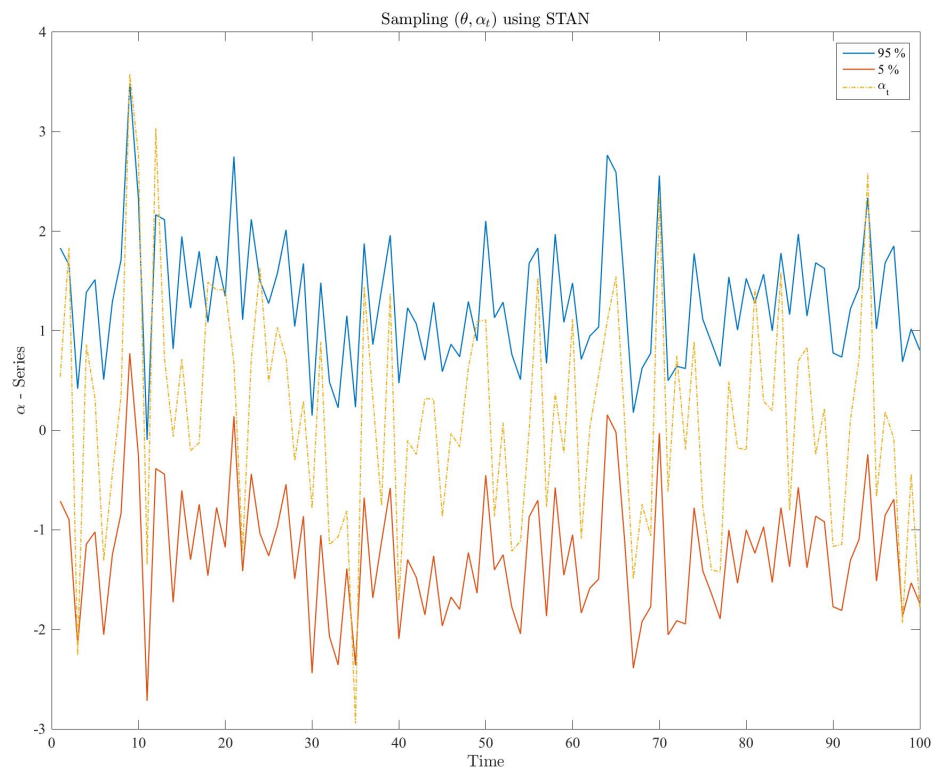
```
tic
launch_stan_full
time_stan_full = toc
```

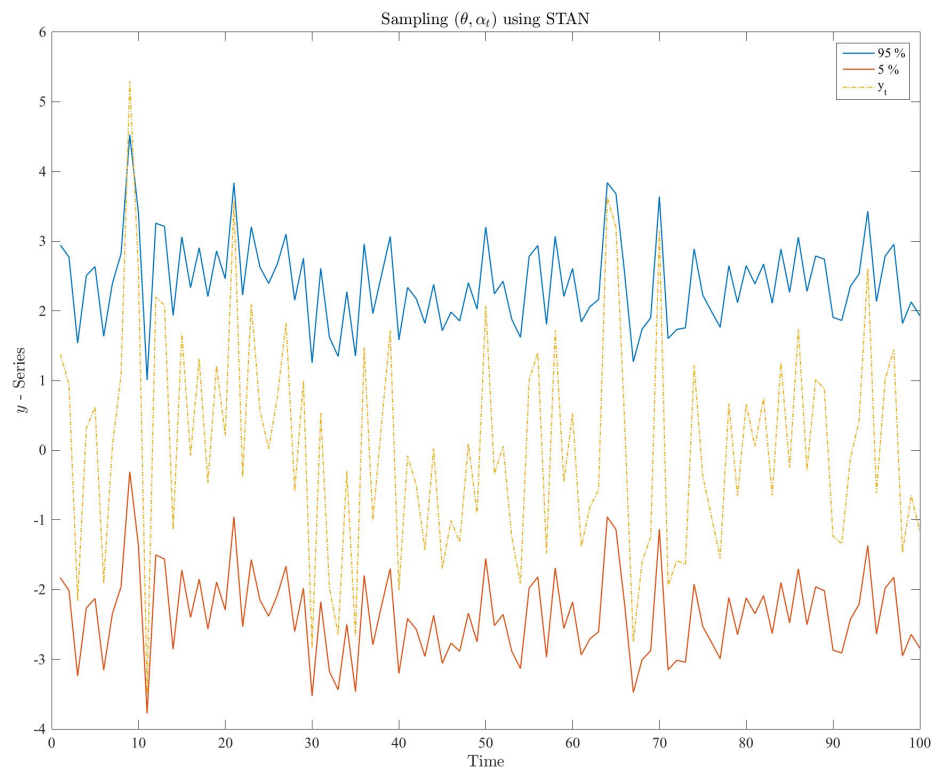
Stan succeeded.

```
time_stan_full =  
36.7566
```







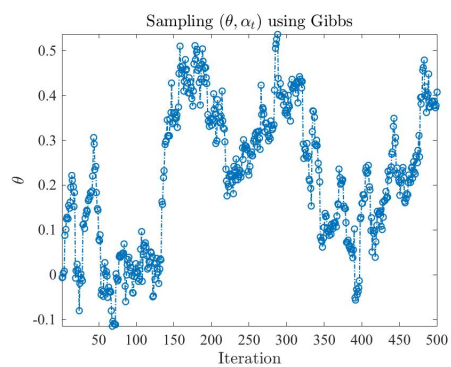
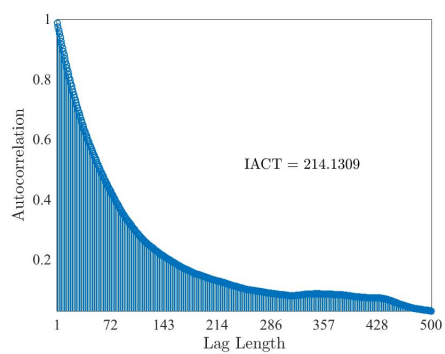
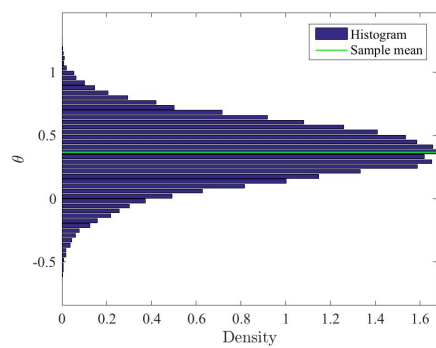
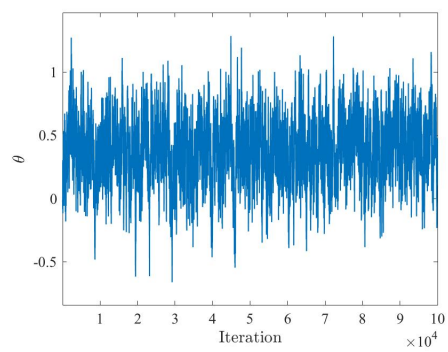


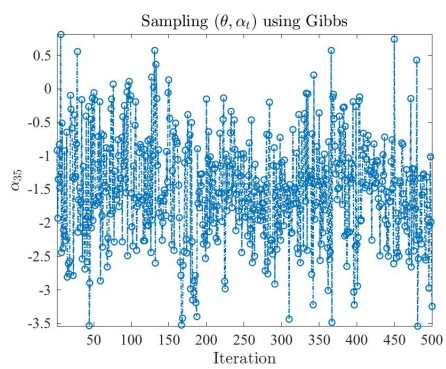
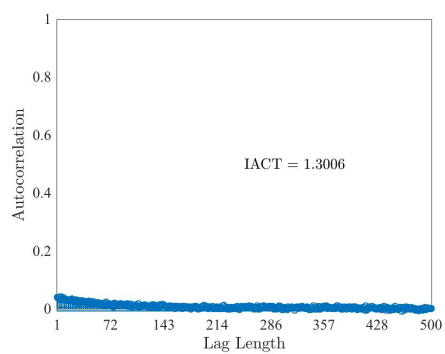
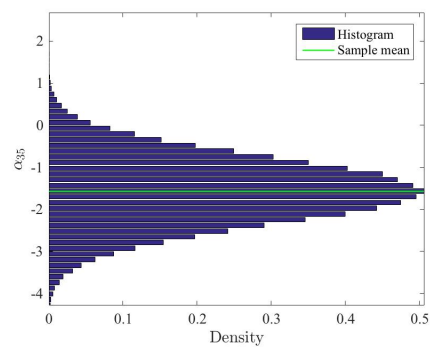
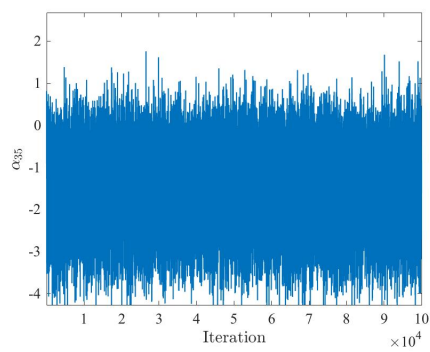
Sampling (θ, α_t) using T-factors Auxiliary Gibbs sampler

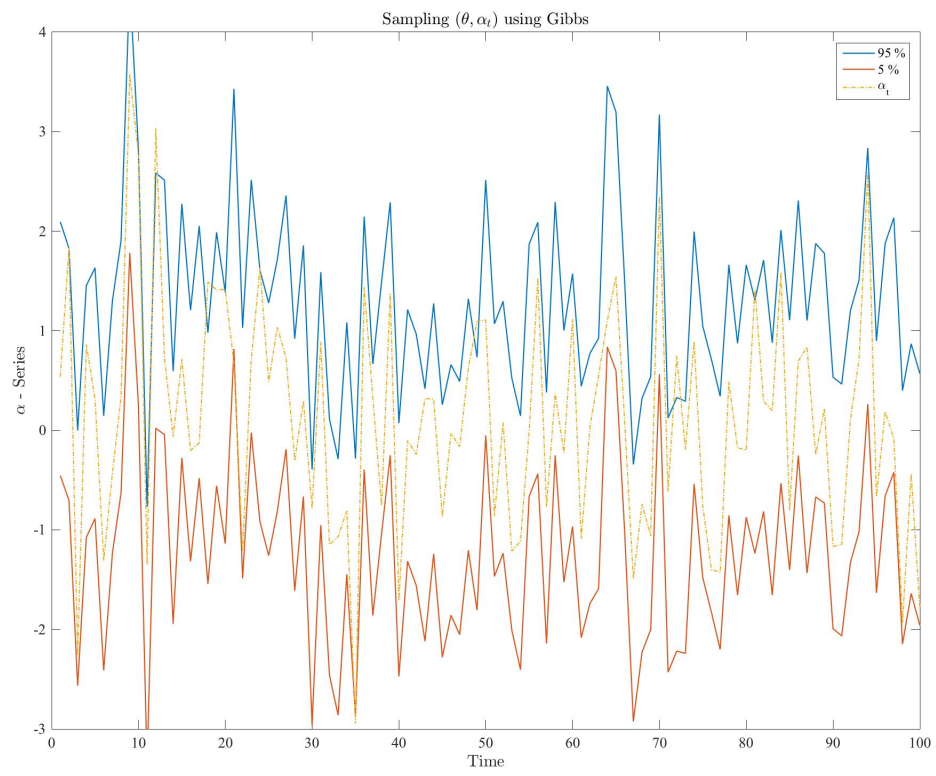
```
tic  
T_factors_Gibbs  
time_Gibbs = toc
```

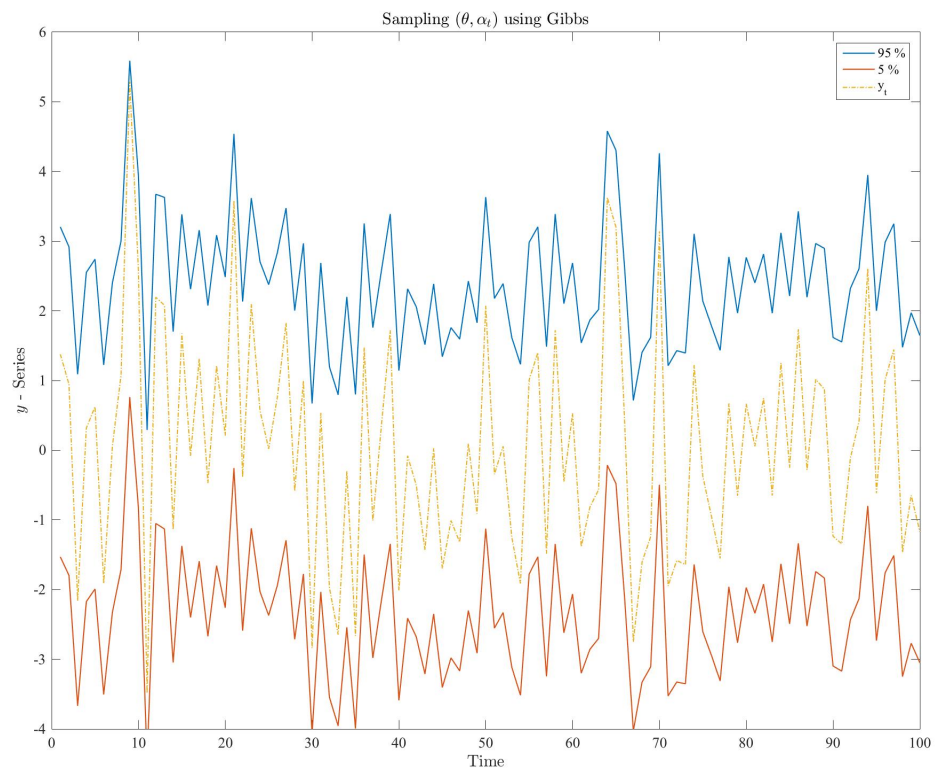
```
time_Gibbs =
```

```
159.8191
```





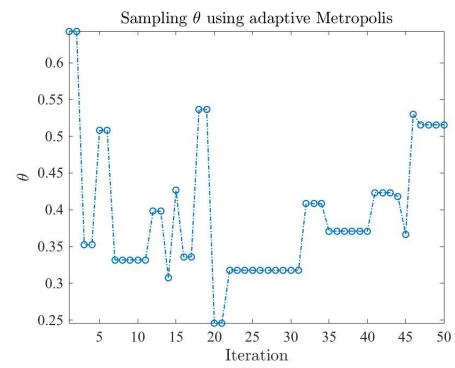
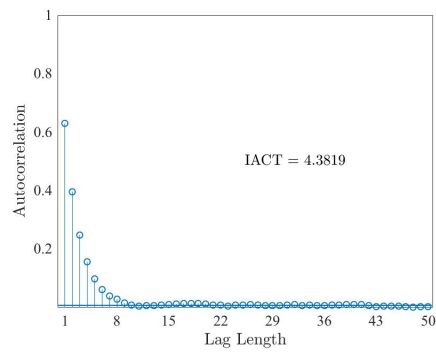
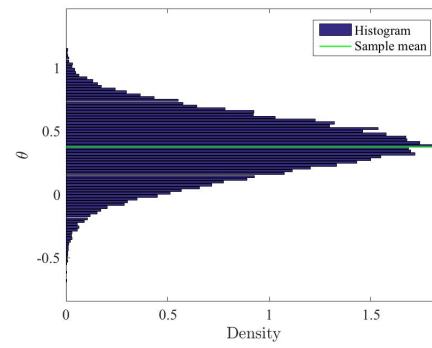
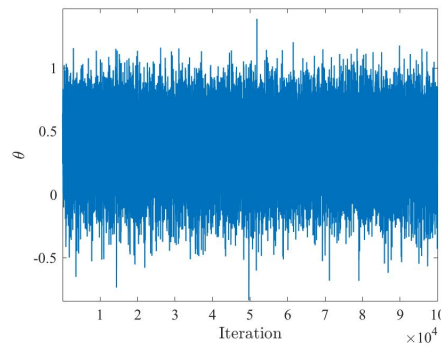




Sampling θ using adaptive Metropolis sampler

```
tic  
launch_am  
time_am = toc
```

```
time_am =  
  
7.9208
```



Cleaning up working folder

```
delete *.stan *.hpp *.mat *.csv *.R
```

Published with MATLAB® R2015b